

3D Medical Image Processing Laboratory



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Center



Factor Analysis in Nuclear Medicine - Useful but Dangerous

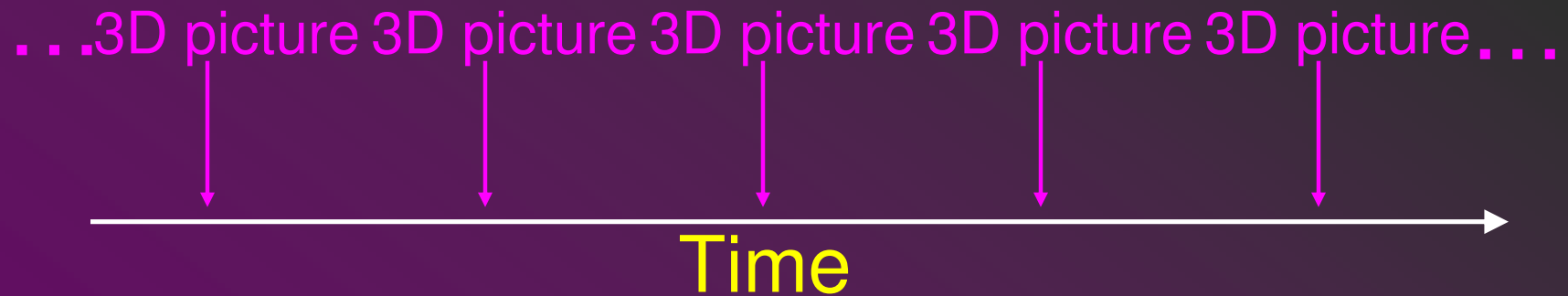
- Introduction
 - What is the factor analysis and where can it be applied
- Part I
 - Why dangerous
- Part II
 - Why useful

Goal

- **Familiarize**
- **Caution**
- **Interest**

Dynamic Imaging

- 4 Dimensions



Dynamic Imaging

- Series of 3D-images each taken at a different time
- Differential uptake of pharmaceutical can be obtained

Factor Analysis in Dynamic Imaging

- Series of 3D-images each taken at a different time

↓
A method of obtaining of
the differential time
uptake for a given organ
or a region in 3D image

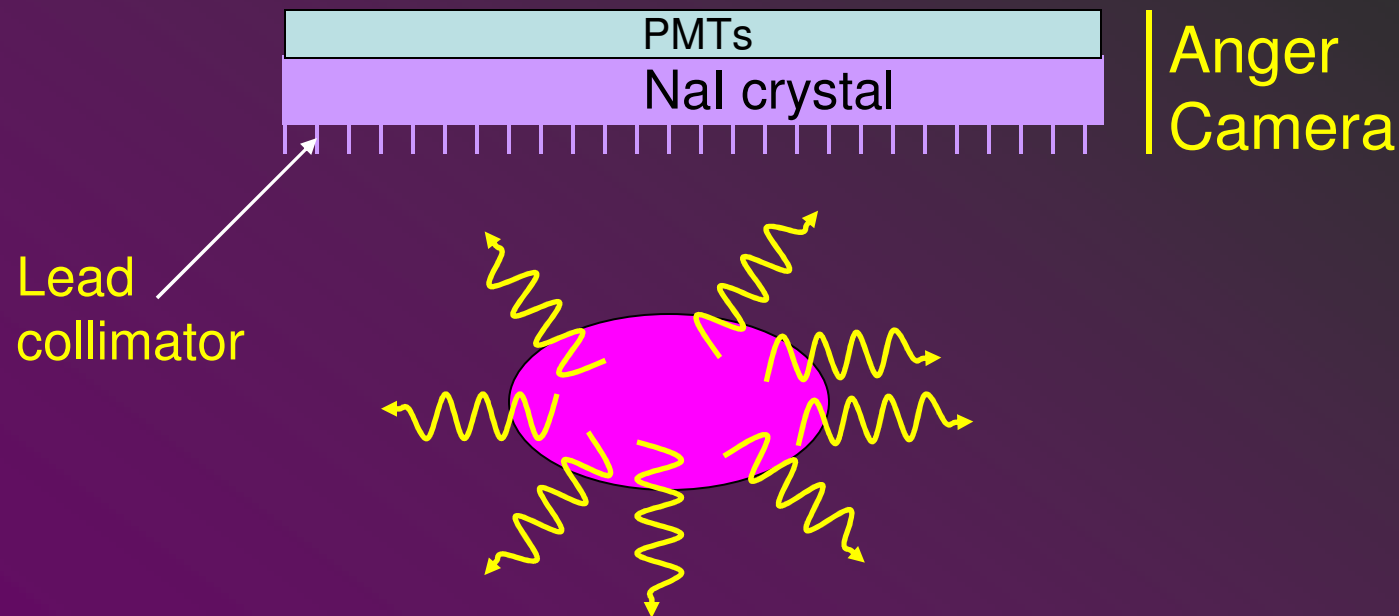
- Differential time uptake of pharmaceutical can be obtained

Examples of Dynamic Imaging

- Planar Single Photon
 ^{99m}Tc -MAG3 for estimating kidney GFR (2Dimensions)
- SPECT
 ^{201}Tl , ^{99m}Tc -Teboroxime for heart perfusion
- PET
 ^{15}O -Water for brain activation study
- PET
 ^{11}C -Raclopride ^{18}F -DMFP for dopamine studies in a brain
- PET
 ^{13}N -Ammonia, ^{82}Ru , ^{18}F -FDG, ^{18}F -fatty acids for cardiac studies

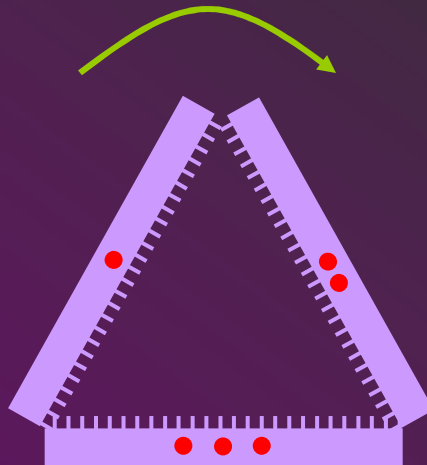
Acquisition of Dynamic Data

- Planar
 - 2D images
 - Stationary detector



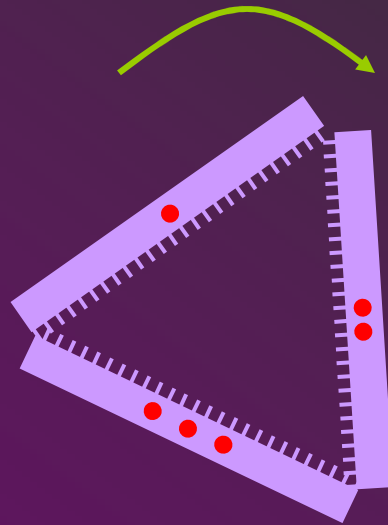
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



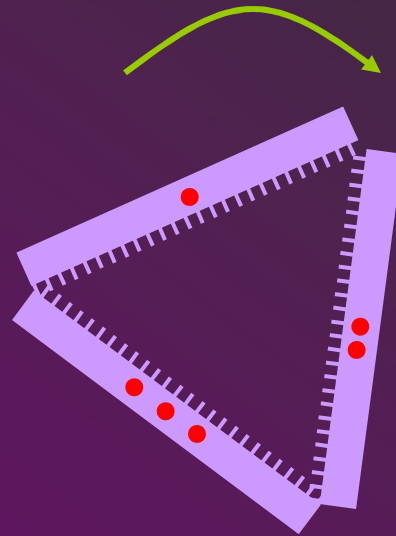
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



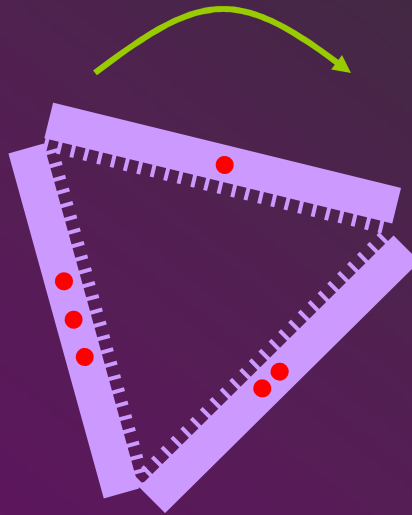
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



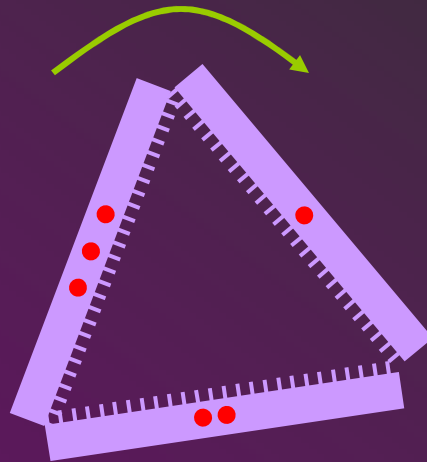
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



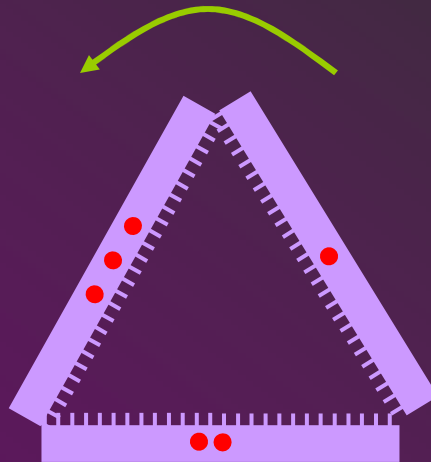
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



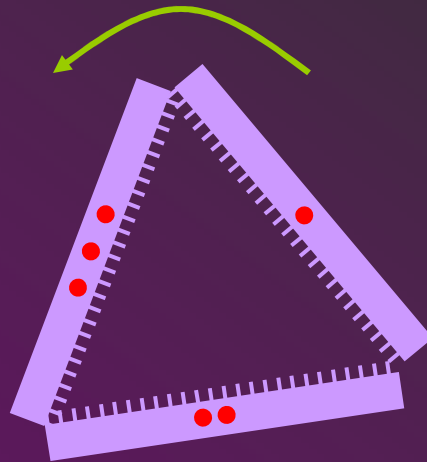
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



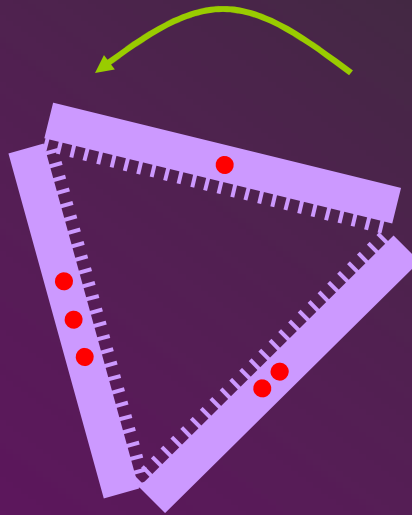
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



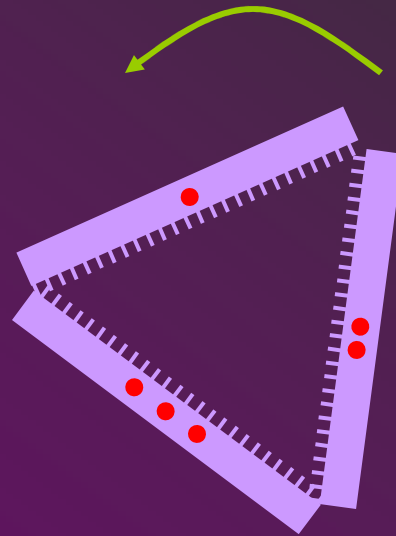
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



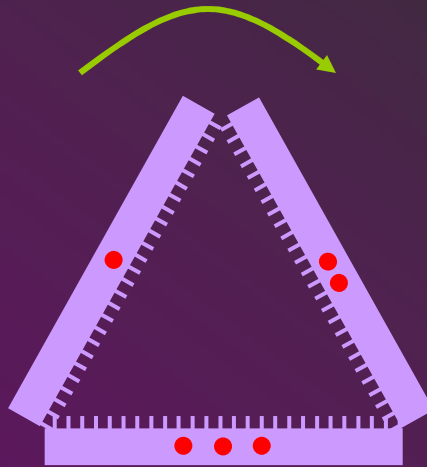
Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



Acquisition of Dynamic Data

- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



Acquisition of Dynamic Data

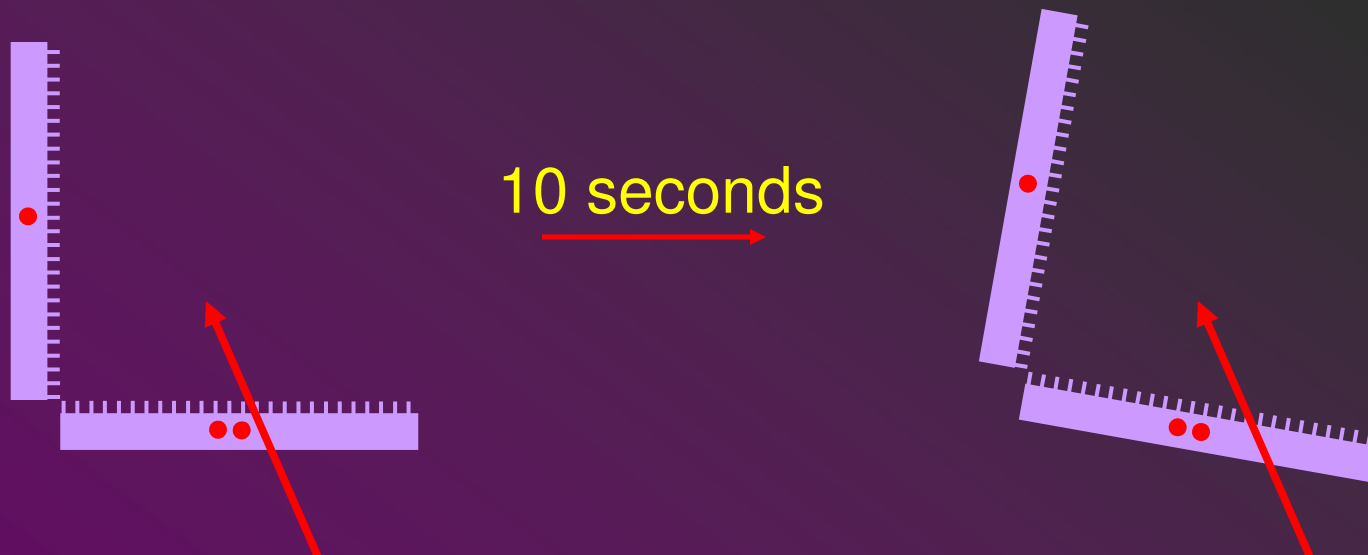
- FAST SPECT
 - Fast rotation
 - Usually done with 3 detector camera



- Each 120° rotation gives one 3D image
- The best time resolution of such system: about 10 seconds

Acquisition of Dynamic Data

- SLOW SPECT
 - Slow rotation
 - Inconsistent projections



- Detectors during this projection see different activity than during this one
- Standard reconstruction of such inconsistent data will FAIL

Acquisition of Dynamic Data

- SLOW SPECT
 - Slow rotation
 - Inconsistent projections
 - Special reconstruction techniques have to be used to deal with the inconsistency in the Slow SPECT acquired dynamic data
 - Factor analysis can be used (Sitek *et al* JNM 2001)

Acquisition of Dynamic Data

- PET

Dynamic PET

EASY

compared to SPECT

“... and sometimes I dream about inherent dynamic PET acquisitions...”

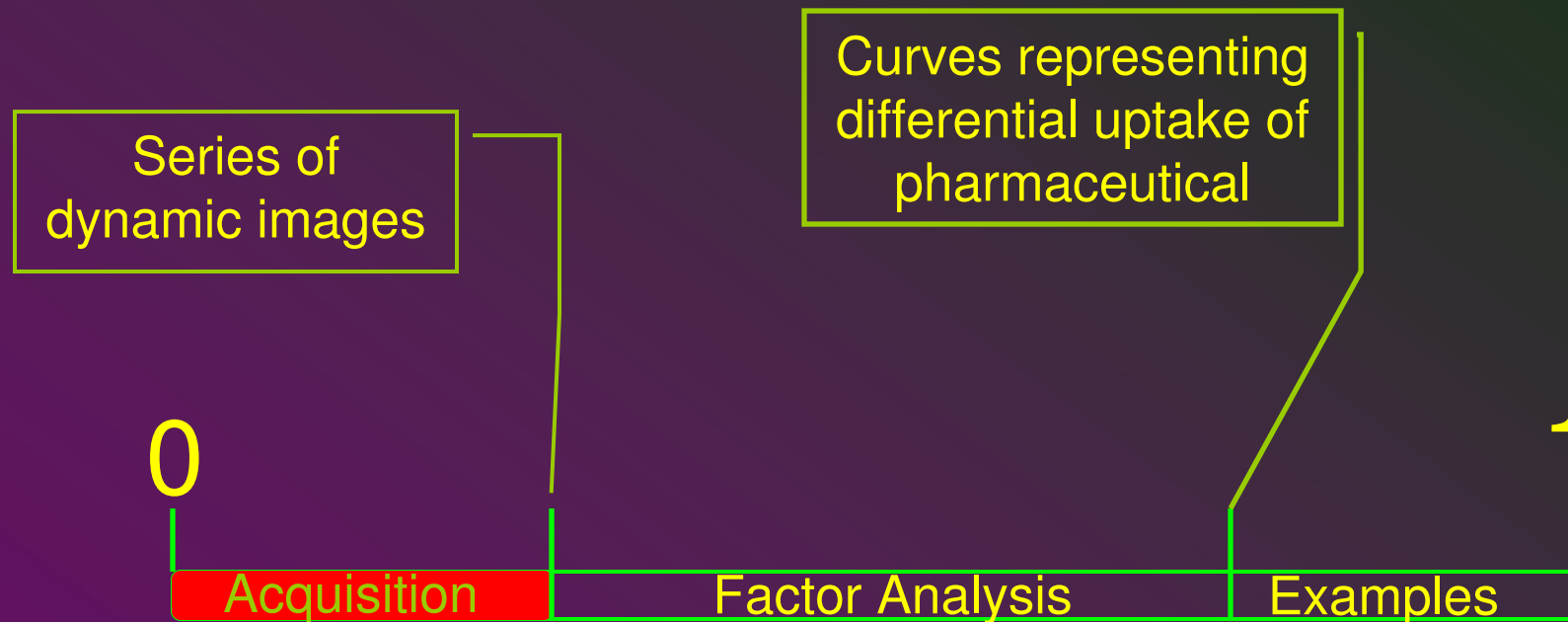
taken from “My Life as a Nuclear Medicine Physicist”

Acquisition of Dynamic Data

- PET

- Projections are consistent - ALWAYS
- Theoretically any time resolution is possible

Presentation Gauge*



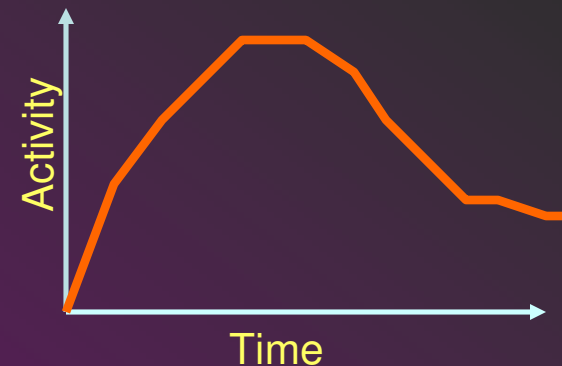
*- any correlation with the time of the presentation should not be assumed

Factor Analysis

- Terminology

Factor Analysis = Factor Analysis of Dynamic Structures (FADS)

Time Activity Curve (TAC)



Factor Analysis

- Factor model assumes that in the image there are groups of pixels that have the same time behavior
- Examples: Liver, heart, heart defect *etc.*

Factor Analysis

- Factor Analysis finds groups of pixels that have a similar time uptake, and also determines the corresponding TAC
- Number of groups is predetermined (usually up to 4)

A factor is the TAC that corresponds to a group

A factor coefficient image is a spatial definition of a group

Factor Analysis

Image from dynamic sequence acquired at time t = Image of factor coefficients for first factor times value of that factor for time t + Image of factor coefficients for second factor times value of that factor for time t + ...

$$I(t) = C_1 F_1(t) + C_2 F_2(t) + \dots$$

$$I = C * F$$

Factor Analysis

Dynamic sequence = Factor coefficients * Factor

$$\begin{matrix} & M \\ N & \boxed{\mathbf{I}} \end{matrix} = \begin{matrix} & K \\ N & \boxed{\mathbf{C}} \end{matrix} * \mathbf{F} \begin{matrix} & M \\ K & \boxed{\phantom{\mathbf{I}}} \end{matrix}$$

N Number of pixels in each dynamic image (128x128x30)

M Number of dynamic images (100)

K Number of factors (4)

Factor Analysis

$$\mathbf{I} = \mathbf{C} * \mathbf{F}$$

The goal of performing FADS is to obtain matrix **C** and matrix **F** having as input matrix **I**

Reminder:

Matrix C corresponds to image of the organs with similar uptake (in cardiac imaging it could be heart tissue, left or right blood pool, liver...)

Matrix F corresponds to time behavior of those regions

Matrix I is the measured dynamic sequence

Factor Analysis

$$\mathbf{I} = \mathbf{C} * \mathbf{F}$$

Is it hard to get C and F using above equation ? **Not really**

Reminder:

Matrix C corresponds to image of the organs with similar uptake (in cardiac imaging it could be heart tissue, left or right blood pool, liver...)

Matrix F corresponds to time behavior of those regions

Matrix I is the measured dynamic sequence

Factor Analysis

$$\mathbf{I} = \mathbf{C} * \mathbf{F}$$

The most popular methods of solving the above is the apex seeking method by *Di Paola et al IEEE TNS 1980 + Variations on a Theme..*

Reminder:

Matrix C corresponds to image of the organs with similar uptake (in cardiac imaging it could be heart tissue, left or right blood pool, liver...)

Matrix F corresponds to time behavior of those regions

Matrix I is the measured dynamic sequence

Factor Analysis

$$\mathbf{I} = \mathbf{C} * \mathbf{F}$$

Dangerous Point: The solution that come out of these (apex seeking) methods is mathematically not unique if only non-negativity constraints are used

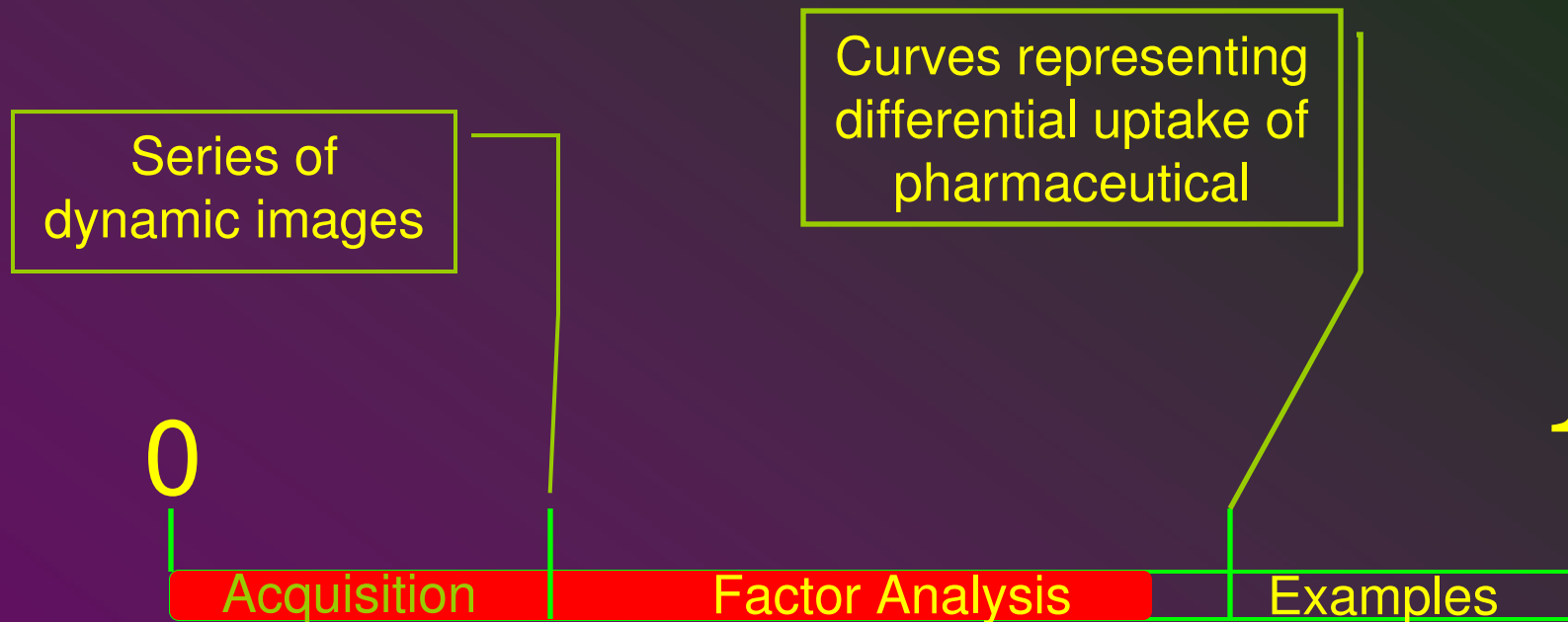
Reminder:

Matrix C corresponds to image of the organs with similar uptake (in cardiac imaging it could be heart tissue, left or right blood pool, liver...)

Matrix F corresponds to time behavior of those regions

Matrix I is the measured dynamic sequence

Presentation Gauge*



*- any correlation with the time of the presentation should not be assumed

Factor Analysis

Demonstration of the non-uniqueness for 2 factor model. Matrices C_1, C_2, F_1, F_2 are non-negative

$$I = C_1 F_1 + C_2 F_2 \quad (1)$$

$$I = C_1 F_1 + C_2 F_2 + a C_1 F_2 - a C_1 F_2 \quad (2)$$

$$I = C_1 (F_1 - a F_2) + (C_2 + a C_1) F_2 \quad (3)$$

$$I = C_1 F'_1 + C'_2 F_2 \quad (4)$$

Matrices $C_1, C'_2 = C_2 + a C_1, F'_1 = F_1 - a F_2, F_2$ are equally good as long as non-negative

Factor Analysis

$$\mathbf{I} = \mathbf{C} * \mathbf{F}$$

In all examples that will be presented factor analysis based on least squares (Sitek *et al* *PMB* 2000, *IEEE TMI* 2002) will be used for solving the above

Reminder:

Matrix C corresponds to image of the organs with similar uptake (in cardiac imaging it could be heart tissue, left or right blood pool, liver...)

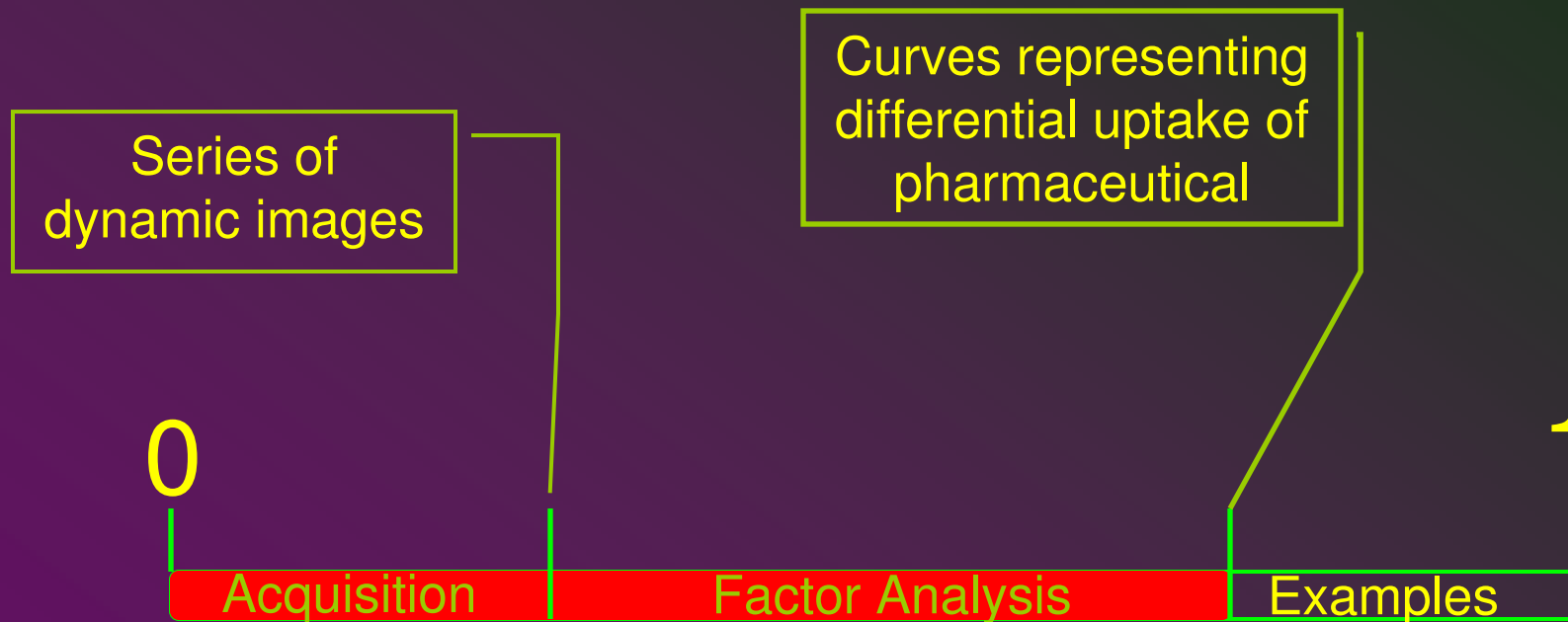
Matrix F corresponds to time behavior of those regions

Matrix I is the measured dynamic sequence

Factor Analysis

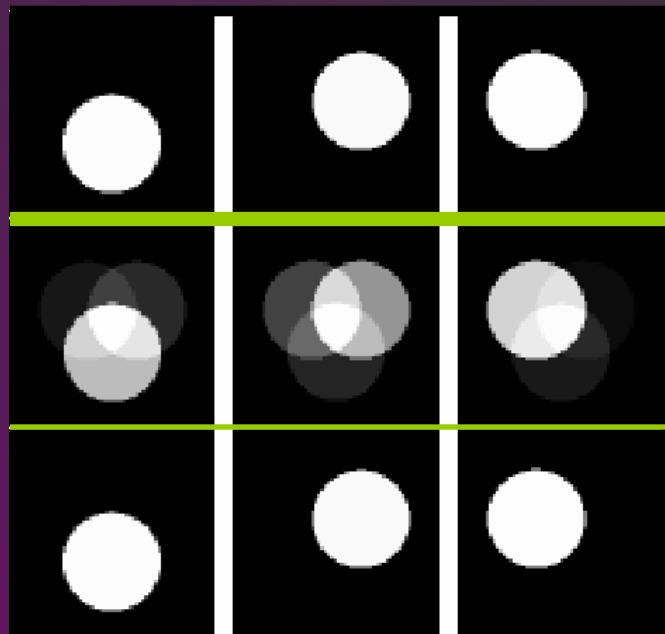
- **LS-FADS** method minimizes the difference between the model and the data with non-negativity constraints (results similar to the apex-seeking)
- **PLS-FADS** method minimizes the difference between the model and the data with the non-negativity constraints and with the correction for non-uniqueness

Presentation Gauge*



*- any correlation with the time of the presentation should not be assumed

FADS examples



Original

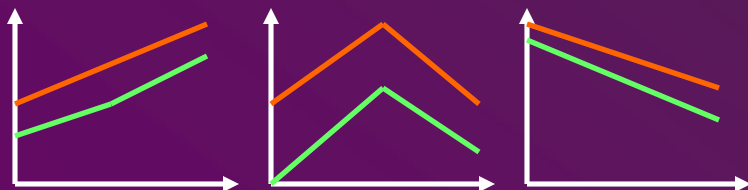
LS-FADS

PLS-FADS

Computer
Simulation



Original or PLS-FADS



LS-FADS

FADS examples

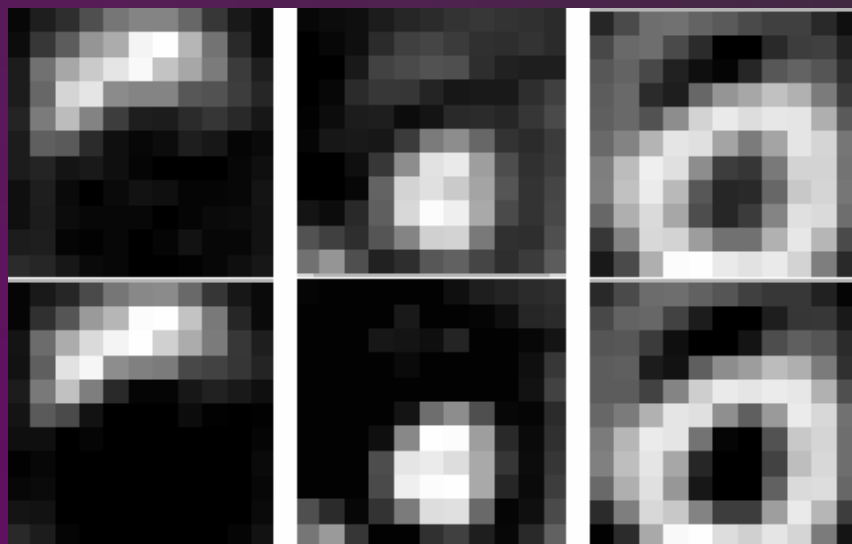
Cardiac canine

^{99m}Tc -Teboroxime

RV

LV

Tissue



LS-FADS

PLS-FADS

IMAGING PROTOCOL

FAST SPECT 179 image dynamic series. Images taken every 7 seconds at rest and stress

Factor Analysis

Demonstration of the non-uniqueness for 2 factor model. Matrices C_1, C_2, F_1, F_2 are non-negative

$$I = C_1 F_1 + C_2 F_2$$

$$I = C_1 F_1 + C_2 F_2 + a C_1 F_2 - a C_1 F_2$$

$$I = C_1 (F_1 - a F_2) + (C_2 + a C_1) F_2$$

$$I = C_1 F'_1 + C'_2 F_2$$

Matrices $C_1, C'_2 = C_2 + a C_1, F'_1 = F_1 - a F_2, F_2$ are equally good as long as non-negative

FADS examples

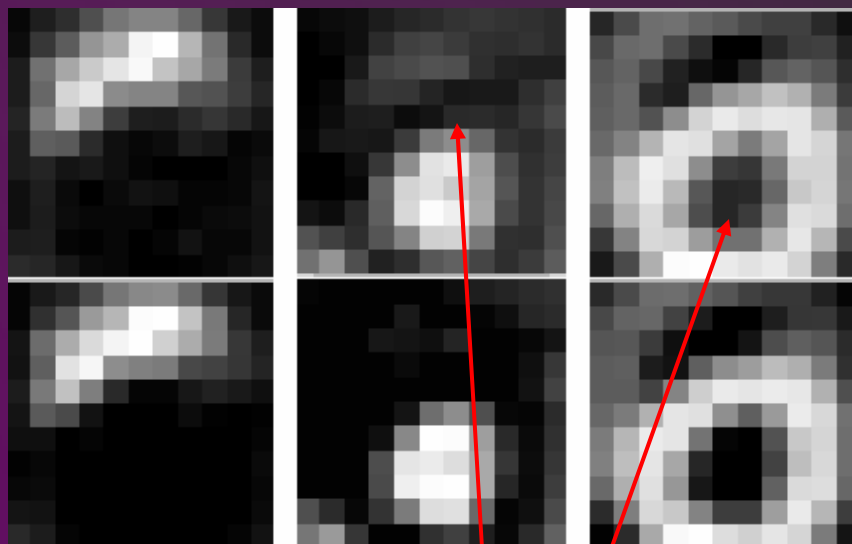
Cardiac canine

^{99m}Tc -Teboroxime

RV

LV

Tissue



LS-FADS

PLS-FADS

Decreased contrast

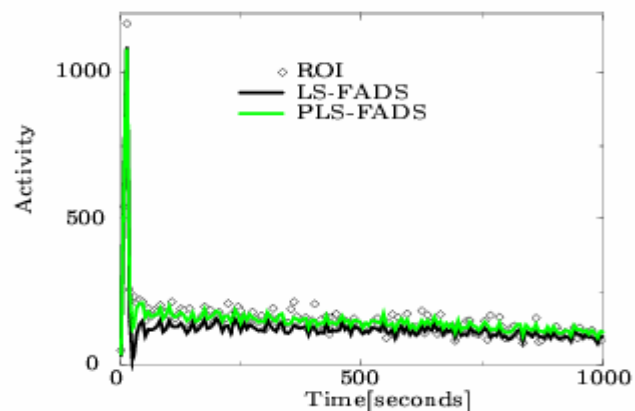
FADS examples

RV

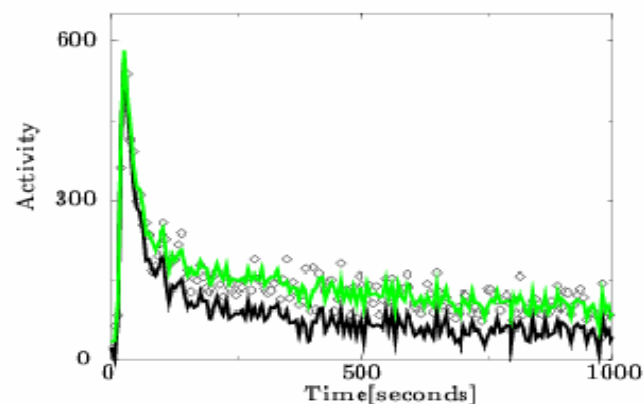
Cardiac canine

^{99m}Tc -Teboroxime

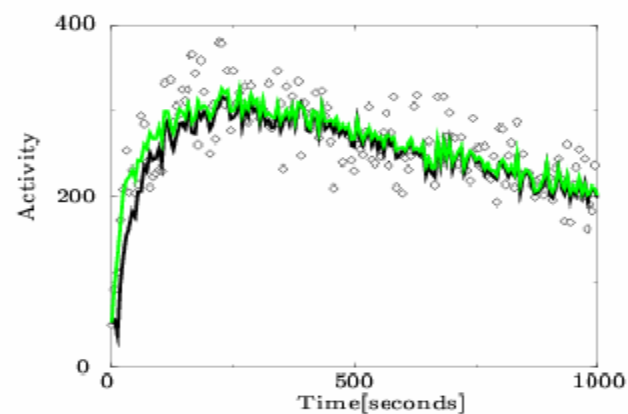
(A)



(B)



(C)



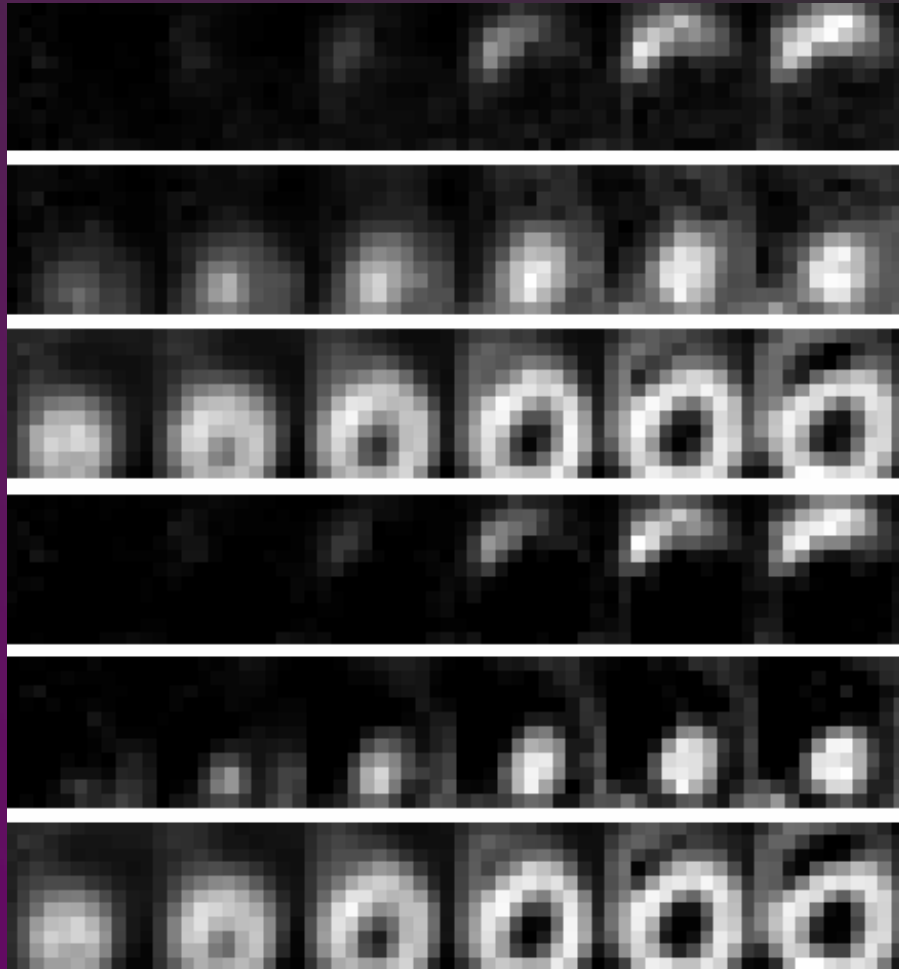
LV

Tissue

FADS examples

Cardiac canine

^{99m}Tc -Teboroxime



RV

LV

Tissue

RV

LV

Tissue

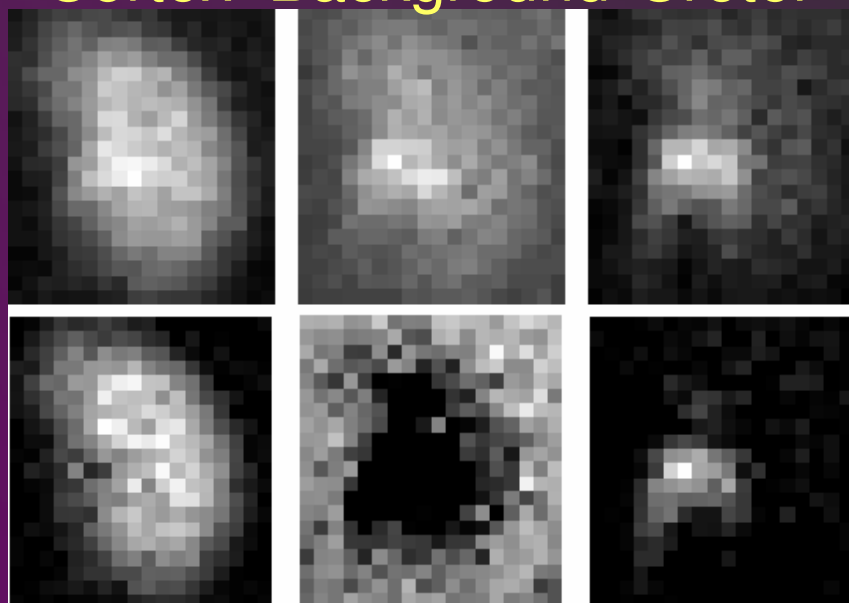
3D

FADS examples

Kidney

^{99m}Tc -MAG3

Cortex Background Ureter



LS-FADS

PLS-FADS

IMAGING PROTOCOL

Planar - 300 image dynamic series.
Images taken every 5 seconds.

FADS examples

LS-FADS PLS-FADS

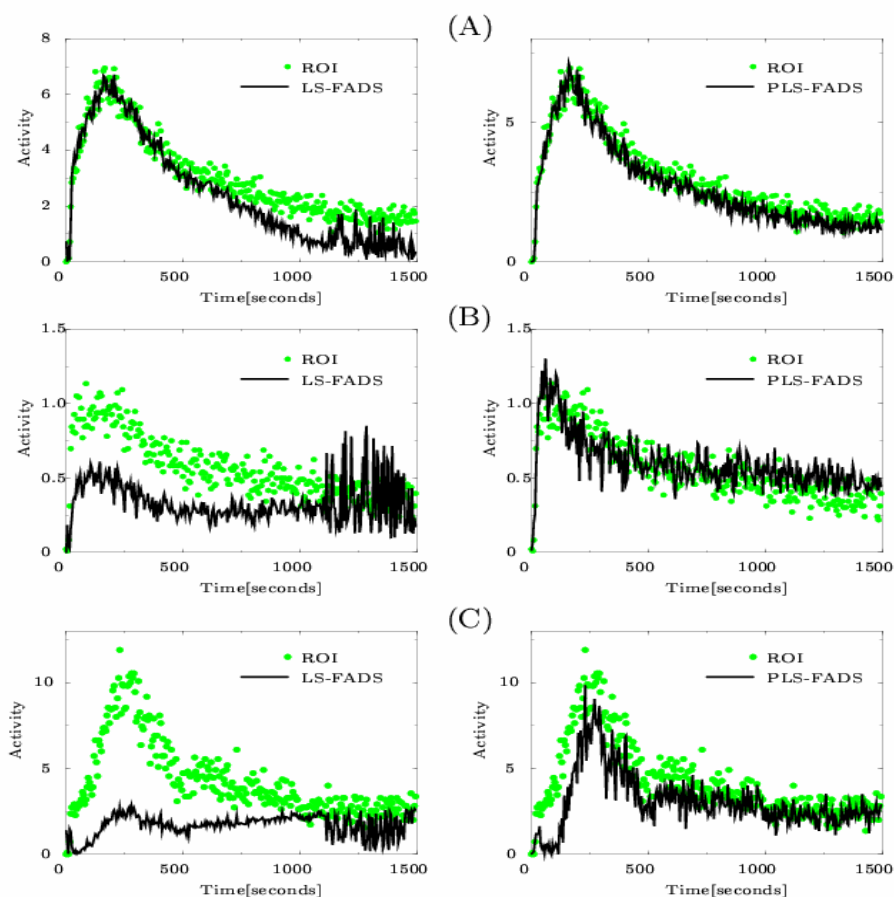
Patient kidney

^{99m}Tc -MAG3

Cortex

Background

Ureter



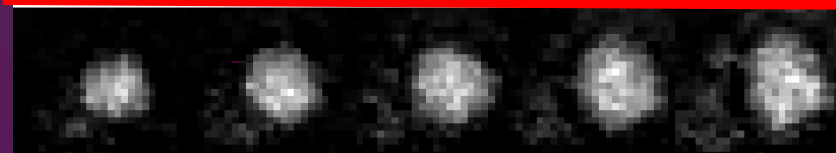
FADS examples

- With no non-uniqueness correction – 2-3 factors
- If the non-uniqueness is properly addressed even 4 factor-FADS works ! Wow !

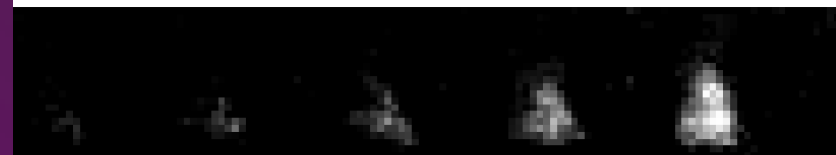
FADS examples



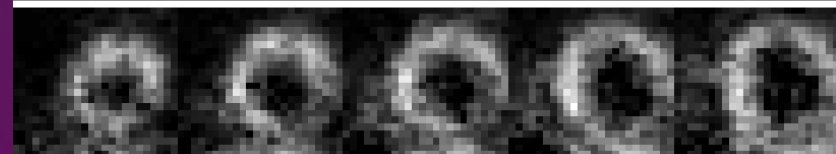
Summed



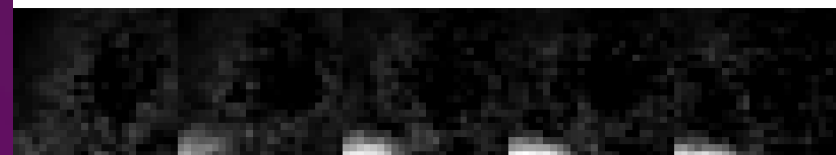
LV



RV



Tissue



Liver



Summed - Liver

Patient cardiac

^{99m}Tc -Teboroxime

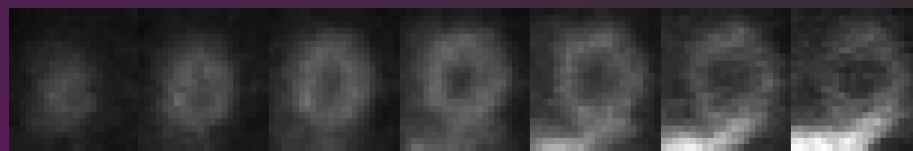
IMAGING PROTOCOL

FAST SPECT - 90 image dynamic series. Images taken every 11 seconds in rest and stress

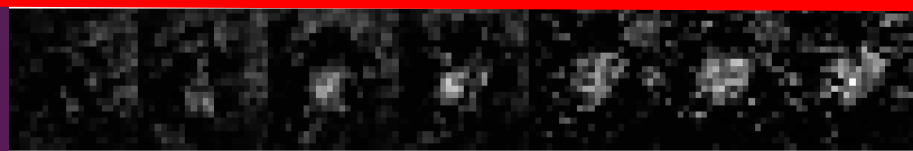
PLS-FADS

FADS examples

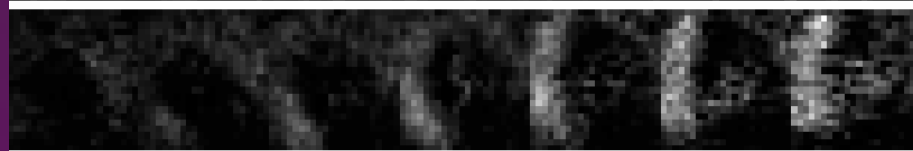
Patient cardiac
 ^{99m}Tc -Teboroxime



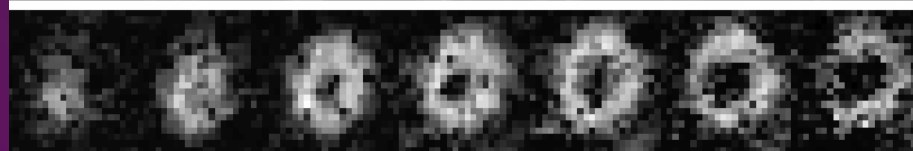
Summed



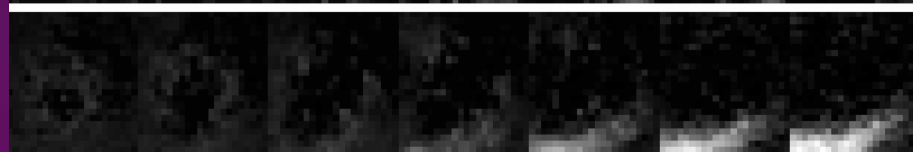
LV



RV



Tissue



Liver



Summed - Liver

PLS-FADS

FADS examples

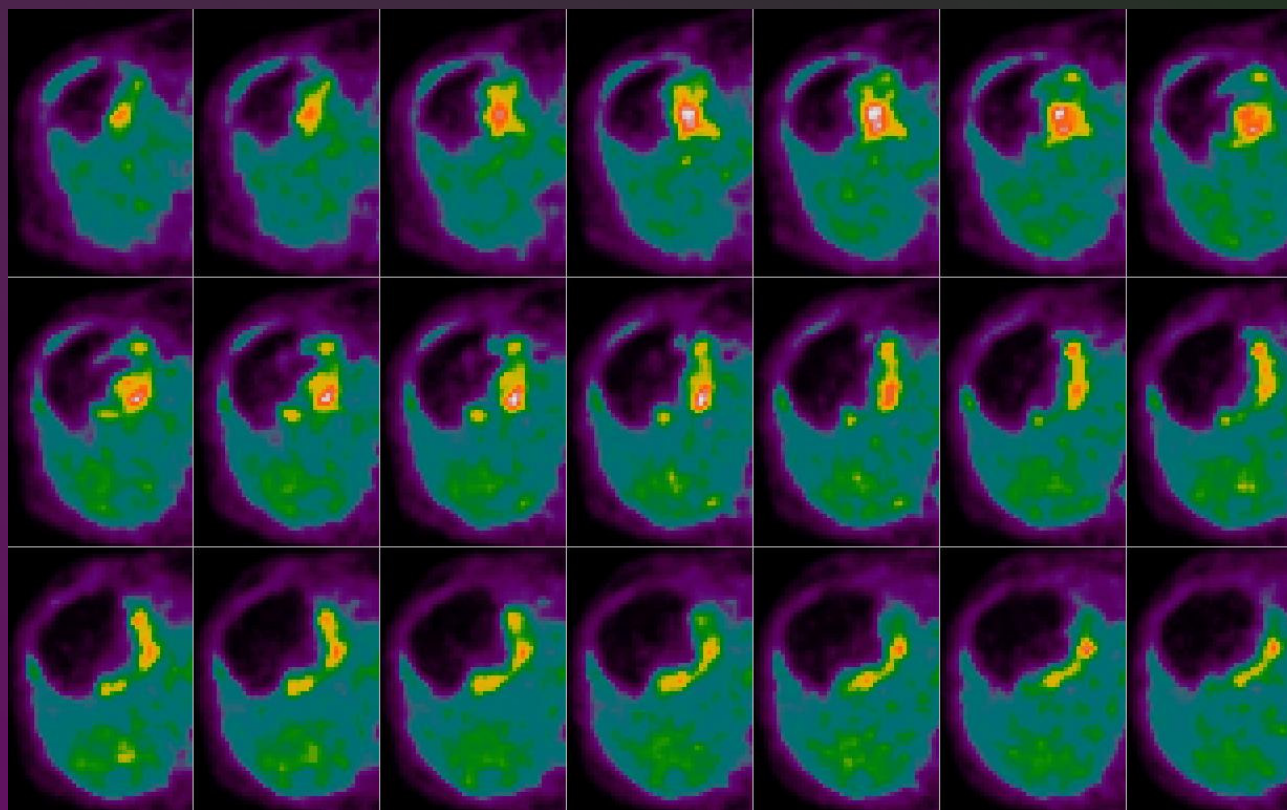
IMAGING PROTOCOL

PET - 30 image dynamic series
Images taken over variable
periods.

Patient liver

^{18}F -FDG

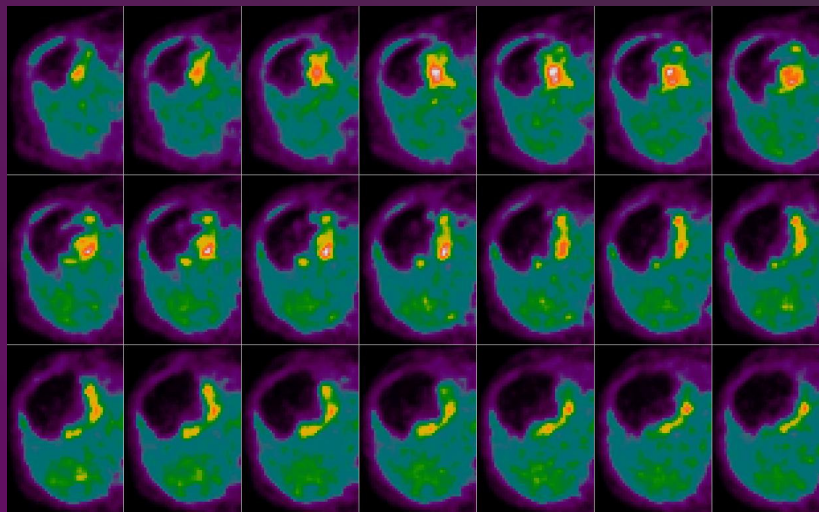
21 slices of
summed image
(right lobe of the
liver)



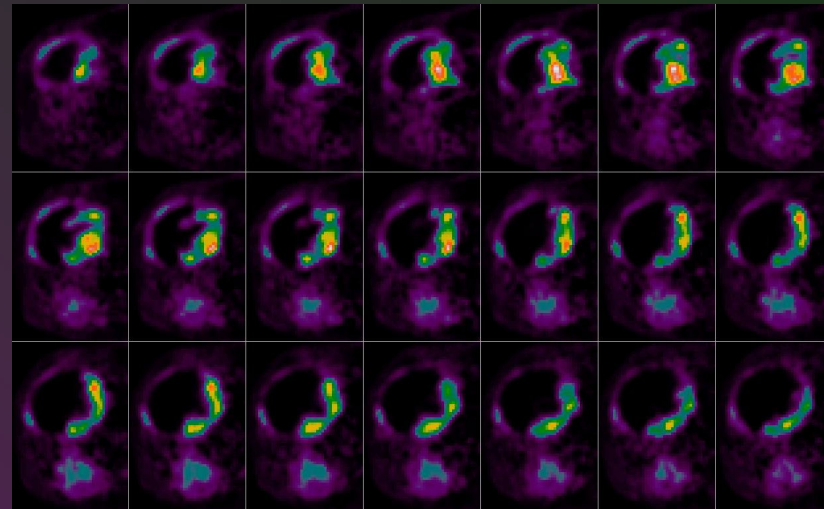
FADS examples

Patient liver ^{18}F -FDG

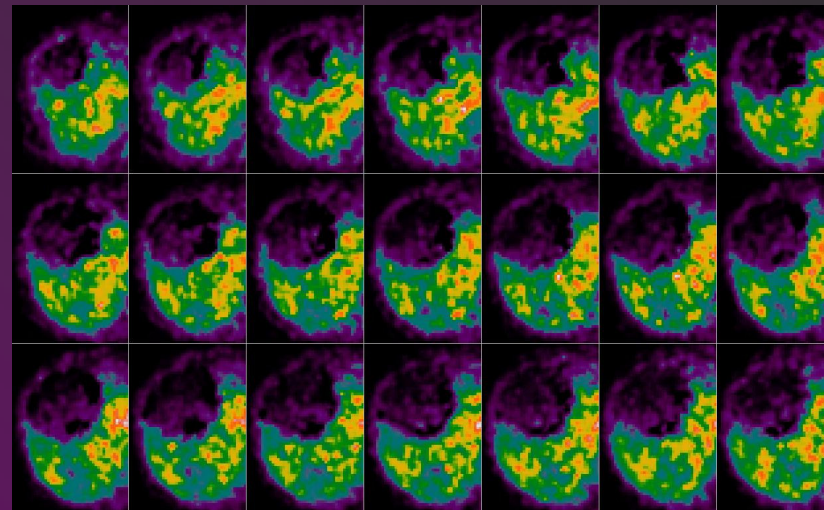
Summed



Tumor



Normal tissue



Summary

- Is this non-uniqueness a problem?
- Does that mean that the factor analysis used in the past 20 years is all wrong?
- Can the non-uniqueness be corrected?

Summary

- Does that mean that the factor analysis with non-negativity constraints used in the past 20 years is all wrong?

YES



These factors create horrible non-uniqueness problems

MAYBE



These factors create some or none non-uniqueness problems

NO



These actually give UNIQUE answer

Summary

- Is this non-uniqueness a problem?
- Does that mean that the factor analysis used in the past 20 years is all wrong?
- Can the non-uniqueness be corrected?

Summary

- Part I
 - Why dangerous

Non-uniqueness

- Part II
 - Why useful

Summary

- Excellent for image segmentation and separation of the overlapped regions
- Very good for extraction of TACs (better than ROI measurements)
- Semi-automatic

“... and they are all this only if you know how to avoid their dangers”

taken from “My Life as a Nuclear Medicine Physicist”

Arkadiusz Sitek

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Thank you